

AD-A127 941

A CONDITION NUMBER FOR DIFFERENTIABLE CONVEX
INEQUALITIES(U) WISCONSIN UNIV-MADISON MATHEMATICS
RESEARCH CENTER O L MANGASARIAN JAN 83 NRC-TSR-2476
DAAG29-86-C-0041

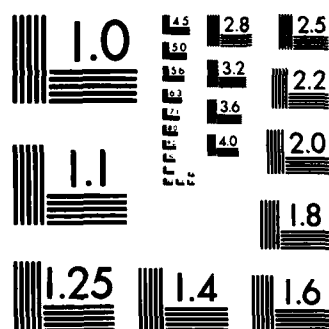
1/1

UNCLASSIFIED

F/G 12/1

NL





MICROCOPY RESOLUTION TEST CHART
NATIONAL BUREAU OF STANDARDS-1963-A

MRC Technical Summary Report #2476

A CONDITION NUMBER FOR
DIFFERENTIABLE CONVEX INEQUALITIES

AD A 127941

O. L. Mangasarian

Mathematics Research Center
University of Wisconsin-Madison
610 Walnut Street
Madison, Wisconsin 53706

January 1983

(Received January 7, 1983)

DTIC
ELECTE
MAY 06 1983
S D
E

Approved for public release
Distribution unlimited

Sponsored by

U. S. Army Research Office
P. O. Box 12211
Research Triangle Park
North Carolina 27709

National Science Foundation
Washington, DC 20550

83 05 06-141

UNIVERSITY OF WISCONSIN-MADISON
MATHEMATICS RESEARCH CENTER

A CONDITION NUMBER FOR DIFFERENTIABLE CONVEX INEQUALITIES

O. L. Mangasarian

Technical Summary Report # 2476

January 1983

ABSTRACT

For a system of differentiable convex inequalities, a new bound is given for the absolute error in an infeasible point in terms of the absolute residual. By using this bound a condition number is defined for the system of inequalities which gives a bound for the relative error in an infeasible point in terms of the relative residual.

AMS (MOS) Subject Classifications: 90C25, 65F35

Key Words: Convex programming, inequalities, condition number

Work Unit Number 5: Mathematical Programming

Sponsored by the United States Army under Contract No. DAAG29-80-C-0041.
This material is based on work supported by National Science Foundation
Grant MCS-8200632.

SIGNIFICANCE AND EXPLANATION

Many important problems can be reduced to finding a feasible point to a system of inequalities. When a feasible point is not readily available or when only an "approximate" feasible point is available, one is interested in determining the error in a given infeasible point. In this work we show how such error can be bounded.

Accession For		
NTIS GRA&I	<input checked="" type="checkbox"/>	
DTIC TAB	<input type="checkbox"/>	
Unannounced	<input type="checkbox"/>	
Justification		
By _____		
Distribution/		
Availability Codes		
Dist	Avail and/or Special	
A		



The responsibility for the wording and views expressed in this descriptive summary lies with MRC, and not with the author of this report.

A CONDITION NUMBER FOR DIFFERENTIABLE CONVEX INEQUALITIES

O. L. Mangasarian

1. Introduction

For a system of linear equalities

$$(1.1) \quad Ax = b$$

where A is a given $n \times n$ real nonsingular matrix and b is a given nonzero vector in the n -dimensional real Euclidean space R^n , the norm of the inverse $\|A^{-1}\|$ and the condition number $\|A^{-1}\| \|A\|$ provide the following useful bounds for the absolute error $\|x - \bar{x}\|$ in terms of the residual $\|Ax - b\|$, and for the relative error $\frac{\|x - \bar{x}\|}{\|\bar{x}\|}$ in terms of the relative residual $\frac{\|Ax - b\|}{\|b\|}$ [1,6]

$$(1.2) \quad \|x - \bar{x}\| \leq \|A^{-1}\| \|Ax - b\|$$

$$(1.3) \quad \frac{\|x - \bar{x}\|}{\|\bar{x}\|} \leq \|A^{-1}\| \|A\| \frac{\|Ax - b\|}{\|b\|}$$

Here, x is any point in R^n , \bar{x} is the exact solution $A^{-1}b$ and $\|\cdot\|$ denotes a vector norm on R^n or its subordinate matrix norm [1,6]. The condition number $\|A^{-1}\| \|A\|$, which depends on the specific norm employed and which is never less than 1, provides a very useful stability measure for the system (1.1). It is the purpose of this work to obtain a corresponding number for the system of inequalities

$$(1.4) \quad g(x) \leq 0$$

where $g: R^n \rightarrow R^m$ is differentiable and convex on R^n . The key to obtaining a condition number for (1.4) is the definition of a quantity that plays the same role as that of $\|A^{-1}\|$ for (1.1) and which would provide an error bound similar to (1.2). In [2] Hoffman extended the bound (1.2) to a system of linear inequalities, and in [5] a new explicit expression was derived for that bound for a system of linear inequalities and equalities. In [7] Robinson extended the bound (1.2) to a system of convex inequalities that define a bounded feasible region with a nonempty interior. In Section 2 of this paper we shall extend the bound (1.2) to a system of differentiable convex inequalities which satisfy a constraint qualification, but without any boundedness assumption on the feasible region. The diameter of the bounded feasible region which appears linearly in Robinson's bound [7, equation (4)] does not appear in our bound (2.3). In Section 3 of this paper we employ the results of Section 2 to obtain a condition number for (1.4) and thereby extend the relative error bound (1.3) to a system of differentiable convex inequalities satisfying a constraint qualification.

We briefly describe now the notation and some of the basic concepts used in this work. For a vector x in the n -dimensional real Euclidean space R^n , $|x|$ and x_+ will denote the vectors in R^n with components $|x|_i = |x_i|$ and $(x_+)_i = \max\{x_i, 0\}$, $i=1,2,\dots,n$, respectively. For a norm $\|x\|_\beta$ on R^n , $\|x\|_{\beta^*}$ will denote the dual norm on R^n , that is $\|x\|_{\beta^*} = \max_{\|y\|_\beta=1} xy$, where xy denotes the scalar product. The generalized Cauchy-Schwarz inequality $|xy| \leq \|x\|_\beta \|y\|_{\beta^*}$, for x

and y in R^n , follows immediately from this definition of the dual norm. For $1 \leq p, q \leq \infty$ and $\frac{1}{p} + \frac{1}{q} = 1$ the p -norm $(\sum_{i=1}^n |x_i|^p)^{\frac{1}{p}}$ and the q -norm are dual norms on R^n [3]. If $\|\cdot\|_\beta$ is a norm on R^n , we shall, with a slight abuse of notation, let $\|\cdot\|_\beta$ also denote the corresponding norm on R^m for $m \neq n$. For an $m \times n$ real matrix A , A_i denotes the i th row and $A_{\cdot j}$ denotes the j th column, while $\|A\|_\beta$ denotes the matrix norm [1,6] subordinate to the vector norm $\|\cdot\|_\beta$, that is $\|A\|_\beta = \max_{\|x\|_\beta=1} \|Ax\|_\beta$. The consistency condition $\|Ax\|_\beta \leq \|A\|_\beta \|x\|_\beta$ follows immediately from this definition of a matrix norm. We shall also use $\|\cdot\|$ to denote an arbitrary vector norm and its subordinate matrix norm. A vector norm $\|\cdot\|$ on R^n is said to be monotonic if and only if $|x_i| \leq |y_i|, i=1, \dots, n$, implies that $\|x\| \leq \|y\|$ [6]. The p -norms $1 \leq p \leq \infty$, are all monotonic norms [6]. A vector of ones in any real Euclidean space will be denoted by e . For a differentiable function $g: R^n \rightarrow R^m$, $\nabla g(x)$ will denote the $m \times n$ Jacobian matrix evaluated at the point x in R^n . For a subset $I \subset \{1, 2, \dots, m\}$, $g_I(x)$ or $g_{i \in I}(x)$ will denote those components of $g(x)$ such that $i \in I$. Similarly $\nabla g_I(x)$ will denote the rows $(\nabla g(x))_i$ of $\nabla g(x)$ such that $i \in I$.

2. An Absolute Error Bound for Differentiable Convex Inequalities

We shall use the approach of [5] to obtain a bound (2.3) for the absolute error $\|x-p(x)\|_\gamma$, where x is any infeasible point for (1.4) and $p(x)$ is some feasible point for (1.4), in terms of the absolute residual $\|g(x)_+\|_\beta$. The constant $\mu_{\beta\gamma}$ relating these two quantities plays a similar role for the differentiable convex inequalities (1.4) as $\|A^{-1}\|$ does for the linear system (1.1).

2.1 Theorem Let g be a differentiable convex function from R^n into R^m , let S° and S defined by

$$(2.1) \quad S^\circ := \{x | g(x) < 0, x \in R^n\} \subset S := \{x | g(x) \leq 0, x \in R^n\}$$

be nonempty and let g satisfy the following asymptotic constraint qualification

(2.2) { For each sequence of points $\{x^i\} \subset S$ such that $g_I(x^i) = 0$ and $\nabla g_I(x^i) \neq 0$ for $I \subset \{1, 2, \dots, m\}$, each accumulation point $\bar{\nabla} g_I$ of the sequence $\{\nabla g_I(x^i)\}$ when the latter is bounded, and of the subsequence $\{\nabla g_I(x^i) / \|\nabla g_I(x^i)\|\}$ when $\{\nabla g_I(x^i)\}$ is unbounded, satisfies

$$\bar{\nabla} g_I z > 0 \text{ for some } z \in R^n$$

Then for each x in R^n there exists a point $p(x)$ in S such that

$$(2.3) \quad \|x-p(x)\|_\gamma \leq \mu_{\beta\gamma} \|g(x)_+\|_\beta$$

where $\mu_{\beta\gamma}$ is a constant independent of x and defined by

$$(2.4) \quad \mu_{\beta\gamma} := \alpha_{\gamma\infty} \sup_{w,p} \{ \|w\|_{\beta^*} \mid p \in S, w \geq 0, wg(p)=0, \|w \nabla g(p)\|_1 = 1 \}$$

where $\|\cdot\|_{\beta^*}$ is the dual norm to $\|\cdot\|_{\beta}$ and $\alpha_{\gamma\infty}$ is the positive constant relating the γ -norm and ∞ -norm by

$$\|z\|_{\gamma} \leq \alpha_{\gamma\infty} \|z\|_{\infty} \text{ for all } z \in \mathbb{R}^n$$

Proof We note first that $\mu_{\beta\gamma}$ as defined by (2.4) is finite under the asymptotic constraint qualification (2.2). For if not then there exists a sequence $\{(w^i, p^i)\}$ such that $w^i \neq 0$ and $\|w^i\| \rightarrow \infty$, $p^i \in S$, $w^i \geq 0$, $w^i g(p^i) = 0$ and $\|w^i \nabla g(p^i)\|_1 = 1$. Since $w^i g(p^i) = 0$, $w^i \geq 0$, $g(p^i) \leq 0$, it follows that $g_I(p^i) = 0$ and $w_{j \notin I}^i = 0$ for some fixed nonempty subset I of $\{1, \dots, m\}$ and for a subsequence $\{(w^i, p^i)\}_{i \in L}$ of $\{(w^i, p^i)\}$. Hence $\{\|w_I^i\|\}_{i \in L} \rightarrow \infty$, $\|w_I^i \nabla g_I(p^i)\|_{i \in L} = 1$ and $\nabla g_I(p^i) \neq 0$. For the case when $\{\nabla g_I(p^i)\}_{i \in L}$ is bounded the subsequences $\{\nabla g_I(p^i)\}_{i \in L}$ and $\{w_I^i / \|w_I^i\|\}_{i \in L}$ have respective accumulation points $\bar{\nabla} g_I$ and \bar{w}_I . When $\{\nabla g_I(p^i)\}_{i \in L}$ is unbounded the subsequences $\{\nabla g_I(p^i) / \|\nabla g_I(p^i)\|\}_{i \in L}$ and $\{w_I^i / \|w_I^i\|\}_{i \in L}$ have respective accumulation points $\bar{\nabla} g_I$ and \bar{w}_I . In either case, since $\|w_I^i \nabla g_I(p^i)\|_{i \in L} = 1$ and $\{\|w_I^i\|\}_{i \in L} \rightarrow \infty$, it follows that

$$\bar{w}_I \bar{\nabla} g_I = 0, \bar{w}_I \geq 0, \|\bar{w}_I\| = 1$$

This however contradicts the asymptotic constraint qualification (2.2) that $\bar{\nabla} g_I z > 0$ for some z in \mathbb{R}^n . Consequently $\mu_{\beta\gamma}$ is finite.

Now for any x in \mathbb{R}^n not in S define $p(x)$ as the projection of x on S using the ∞ -norm. Hence $p(x)$ and some $\delta(x) > 0$ solve the following convex programming problem

$$(2.5) \quad \min_{p, \delta} \{ \delta \mid -e\delta \leq p-x \leq e\delta, g(p) \leq 0, p \in R^n, \delta \in R \}$$

Note that since S° is nonempty by assumption, the interior of the feasible region of (2.5) is also nonempty and hence $p(x)$, $\delta(x)$ and some $u(x)$, $v(x)$ and $w(x)$ satisfy the following Karush-Kuhn-Tucker conditions [4] for (2.5) where the explicit dependence of p , δ , u , v and w on x has been dropped for simplicity

$$1 - e(u+v) = 0, u - v + w\nabla g(p) = 0, (u, v, w) \geq 0$$

$$-e\delta \leq p - x \leq e\delta, g(p) \leq 0$$

$$wg(p) = 0, u_j(p-x-e\delta)_j = 0, v_j(-e\delta-p+x)_j = 0, j=1, \dots, n$$

From the last two equalities it follows that $u_j v_j = 0$ for $j=1, \dots, n$, because

$$(p-x-e\delta)_j + (-e\delta-p+x)_j = -2\delta < 0$$

Consequently

$$(2.6) \quad p \in S, wg(p) = 0, w \geq 0, \|\nabla g(p)\|_1 = 1$$

Hence

$$\begin{aligned} 0 &< \|x-p\|_\infty = \delta \\ &= \delta + u(p-x-e\delta) + v(-e\delta-p+x) + wg(p) \\ &= x(v-u) + wg(p) - w\nabla g(p)p \\ &= w\nabla g(p)x + wg(p) - w\nabla g(p)p \\ &\leq wg(x) \quad (\text{By convexity of } g \text{ and } w \geq 0) \\ &\leq wg(x)_+ \end{aligned}$$

$$\leq \|w\|_{\beta^*} \|g(x)_+\|_{\beta} \quad (\text{By Cauchy-Schwarz inequality})$$

$$\leq \frac{\mu_{\beta\gamma}}{\alpha_{\gamma\infty}} \|g(x)_+\|_{\beta} \quad (\text{By (2.4) and (2.6)})$$

Hence

$$\|x-p\|_{\gamma} \leq \alpha_{\gamma\infty} \|x-p\|_{\infty} \leq \mu_{\beta\gamma} \|g(x)_+\|_{\beta} \quad \square$$

2.2 Remark The asymptotic constraint qualification (2.2) is merely used as a sufficient condition for the finiteness of $\mu_{\beta\gamma}$ as defined by (2.4). Hence Theorem 2.1 can be stated with the constraint qualification (2.2) replaced by the assumption that the supremum of (2.4) defining $\mu_{\beta\gamma}$ is finite.

We give now a simple example illustrating the above theorem which is not covered by Robinson's result [7, equation (4)] because the feasible region is unbounded.

2.3 Example $S := \{x | x \in \mathbb{R}^2, x_2 \geq e^{x_1}, x_1 \geq 0\}.$

It is easy to verify that the assumptions of Theorem 2.1 are satisfied with $g_1(x) := e^{x_1} - x_2$, $g_2(x) = -x_1$ and that

$$\mu_{\infty\infty} := \sup_{w,p} \{ \|w\|_1 \mid p \in S, w \geq 0, wg(p) = 0, \|w \nabla g(p)\|_1 = 1 \} = 2$$

Consequently for each x in \mathbb{R}^2 there exists a $p(x)$ in S such that

$$\|x-p(x)\|_{\infty} \leq 2 \left\| \begin{pmatrix} (e^{x_1} - x_2)_+ \\ (-x_1)_+ \end{pmatrix} \right\|_{\infty}$$

The bound $\mu_{\infty\infty} = 2$ is sharp here, for take the sequence of points $x_1 = -t$, $x_2 = 1 - 2t$ with t a nonnegative number converging to zero.

Then

$$\frac{\|x-p(x)\|_{\infty}}{\|g(x)_+\|_{\infty}} = \frac{\|-t, -2t\|_{\infty}}{e^{-t}-1+2t} = \frac{2t}{e^{-t}-1+2t}$$

which approaches 2 as t approaches 0.

3. A Relative Error Bound for Differentiable Convex Inequalities

We extend now the relative error bound (1.3) to convex differentiable inequalities by using the error bound (2.3). We will again use the approach of [5] and will need the following simple lemma established there.

3.1 Lemma [5, Lemma 2] Let $\|\cdot\|_\beta$ be a monotonic norm on R^m and let a, b be in R^m . Then $a \leq b$ implies that $\|(a)_+\|_\beta \leq \|b\|_\beta$.

The following is a direct consequence of the above lemma.

3.2 Lemma Let $g: R^n \rightarrow R^m$ be differentiable and convex on R^n and let $\|\cdot\|_\beta$ be a monotonic norm on R^m . Then

$$\|g(0)_+\|_\beta \leq \|\nabla g(0)p\|_\beta \text{ for } g(p) \leq 0$$

Proof By the convexity of g , $0 \geq g(p) \geq g(0) + \nabla g(0)p$. Hence applying Lemma 3.1 to $g(0) \leq -\nabla g(0)p$ we obtain the desired inequality. \square

3.3 Theorem (Condition number bound) Let the assumptions of Theorem 2.1 hold, let $g(0)_+ \neq 0$ and let $\|\cdot\|_\beta$ be a monotonic norm on R^m . Then for each x in R^n there exists a $p(x)$ in S such that

$$(3.1) \quad \frac{\|x-p(x)\|_\beta}{\|p(x)\|_\beta} \leq \mu_{\beta\beta} \|\nabla g(0)\|_\beta \frac{\|g(x)_+\|_\beta}{\|g(0)_+\|_\beta}$$

where $\mu_{\beta\beta}$ is defined by (2.4), and $\mu_{\beta\beta} \|\nabla g(0)\|_\beta$ defines the condition number of (1.4).

Proof For each x in R^n there exists a $p(x)$ in S such that

$$\frac{\|x-p(x)\|_{\beta}}{\|p(x)\|_{\beta}} \leq \mu_{\beta\beta} \frac{\|g(0)_+\|_{\beta}}{\|p(x)\|_{\beta}} \frac{\|g(x)_+\|_{\beta}}{\|g(0)_+\|_{\beta}} \quad (\text{By Theorem 2.1})$$

$$\leq \mu_{\beta\beta} \frac{\|\nabla g(0)p(x)\|_{\beta}}{\|p(x)\|_{\beta}} \frac{\|g(x)_+\|_{\beta}}{\|g(0)_+\|_{\beta}} \quad (\text{By Lemma 3.2})$$

$$\leq \mu_{\beta\beta} \|\nabla g(0)\|_{\beta} \frac{\|g(x)_+\|_{\beta}}{\|g(0)_+\|_{\beta}}$$

□

For Example 2.3, it is easy to verify that $\|\nabla g(0)\|_{\infty} = 2$ and hence the condition number for the example, using the ∞ -norm, is: $\mu_{\infty\infty} \|\nabla g(0)\|_{\infty} = 4$.

Acknowledgements

I am indebted to Stephen M. Robinson for referring me to his paper [7] and to my student T.-H. Shiau for reading the paper and suggesting improvements.

REFERENCES

1. S. D. Conte and C. de Boor: "Elementary numerical analysis", McGraw-Hill, New York, 1980.
2. A. J. Hoffman: "On approximate solutions of systems of linear inequalities", Journal of Research of the National Bureau of Standards 49, 1952, 263-265.
3. A. S. Householder: "The theory of matrices in numerical analysis", Blaisdell Publishing, New York, 1964.
4. O. L. Mangasarian: "Nonlinear programming", McGraw-Hill, New York, 1969.
5. O. L. Mangasarian: "A condition number for linear inequalities and linear programs", in "Methods of Operations Research", Proceedings of 6. Symposium uber Operations Research, Augsburg 7-9 September 1981, G. Bamberg & O. Optiz (eds.), Verlagsgruppe Athenäum/Hain/Scriptor/Hanstein, Königstein 1981, 3-15.
6. J. S. Ortega: "Numerical analysis", Academic Press, New York, 1972.
7. S. M. Robinson: "An application of error bounds for convex programming in a linear space", SIAM Journal on Control and Optimization 13, 1975, 271-273.

REPORT DOCUMENTATION PAGE		READ INSTRUCTIONS BEFORE COMPLETING FORM
1. REPORT NUMBER #2476	2. GOVT ACCESSION NO.	3. RECIPIENT'S CATALOG NUMBER
4. TITLE (and Subtitle) A Condition Number for Differentiable Convex Inequalities		5. TYPE OF REPORT & PERIOD COVERED Summary Report - no specific reporting period
		6. PERFORMING ORG. REPORT NUMBER
7. AUTHOR(s) O. L. Mangasarian		8. CONTRACT OR GRANT NUMBER(s) MCS-8200632 DAAG29-80-C-0041
9. PERFORMING ORGANIZATION NAME AND ADDRESS Mathematics Research Center, University of 610 Walnut Street Madison, Wisconsin 53706		10. PROGRAM ELEMENT, PROJECT, TASK AREA & WORK UNIT NUMBERS Work Unit Number 5 - Mathematical Programming
11. CONTROLLING OFFICE NAME AND ADDRESS See Item 18 below		12. REPORT DATE January 1983
		13. NUMBER OF PAGES 11
14. MONITORING AGENCY NAME & ADDRESS (if different from Controlling Office)		15. SECURITY CLASS. (of this report) UNCLASSIFIED
		15a. DECLASSIFICATION/DOWNGRADING SCHEDULE
16. DISTRIBUTION STATEMENT (of this Report) Approved for public release; distribution unlimited.		
17. DISTRIBUTION STATEMENT (of the abstract entered in Block 20, if different from Report)		
18. SUPPLEMENTARY NOTES U. S. Army Research Office P. O. Box 12211 Research Triangle Park North Carolina 27709 National Science Foundation Washington, DC 20550		
19. KEY WORDS (Continue on reverse side if necessary and identify by block number) Convex programming, inequalities, condition number		
20. ABSTRACT (Continue on reverse side if necessary and identify by block number) For a system of differentiable convex inequalities, a new bound is given for the absolute error in an infeasible point in terms of the absolute residual. By using this bound a condition number is defined for the system of inequalities which gives a bound for the relative error in an infeasible point in terms of the relative residual.		

END

FILMED

6-83

DTIC